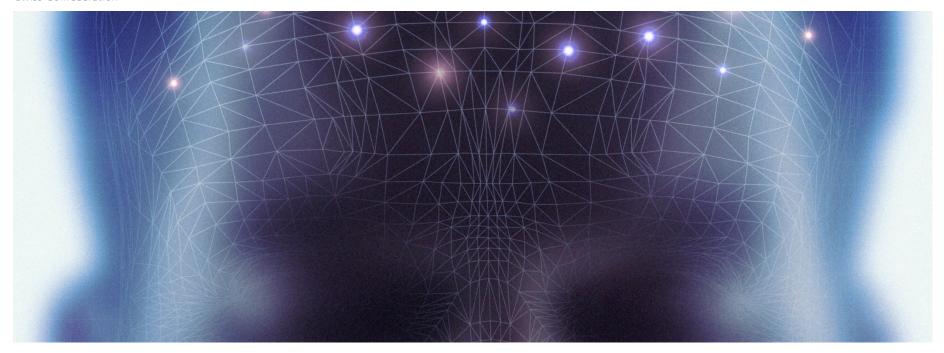


Federal Department of Defence, Civil Protection and Sport DDPS armasuisse Science and Technology

Swiss Confederation

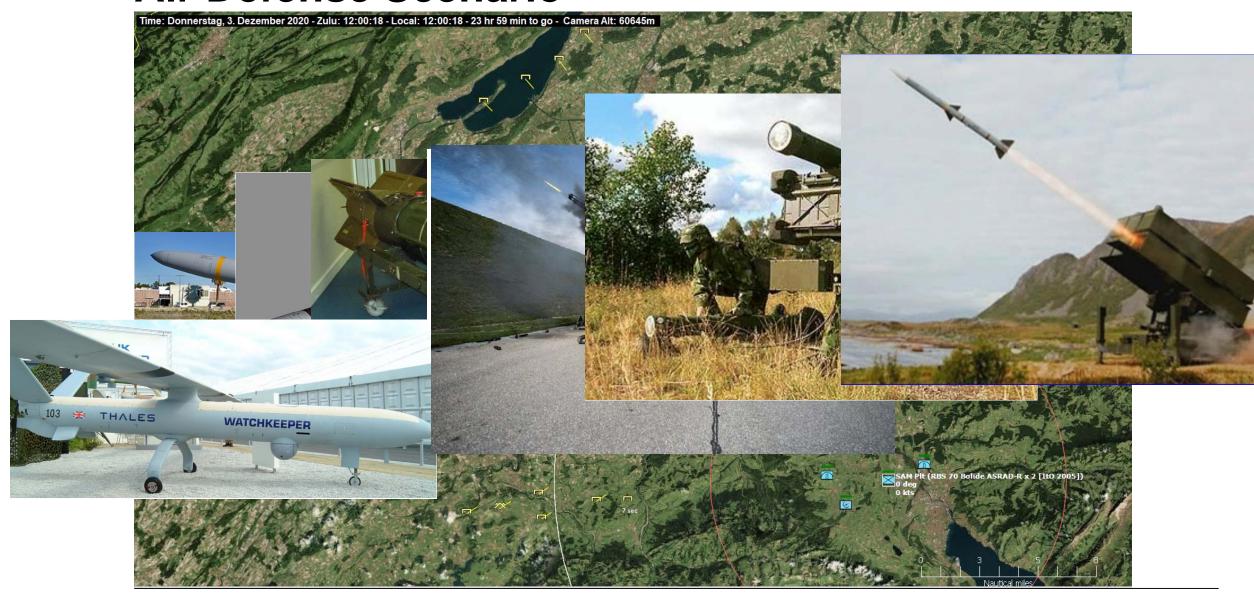


Deep Self-optimizing Artificial Intelligence for Tactical Analysis, Training and Optimization

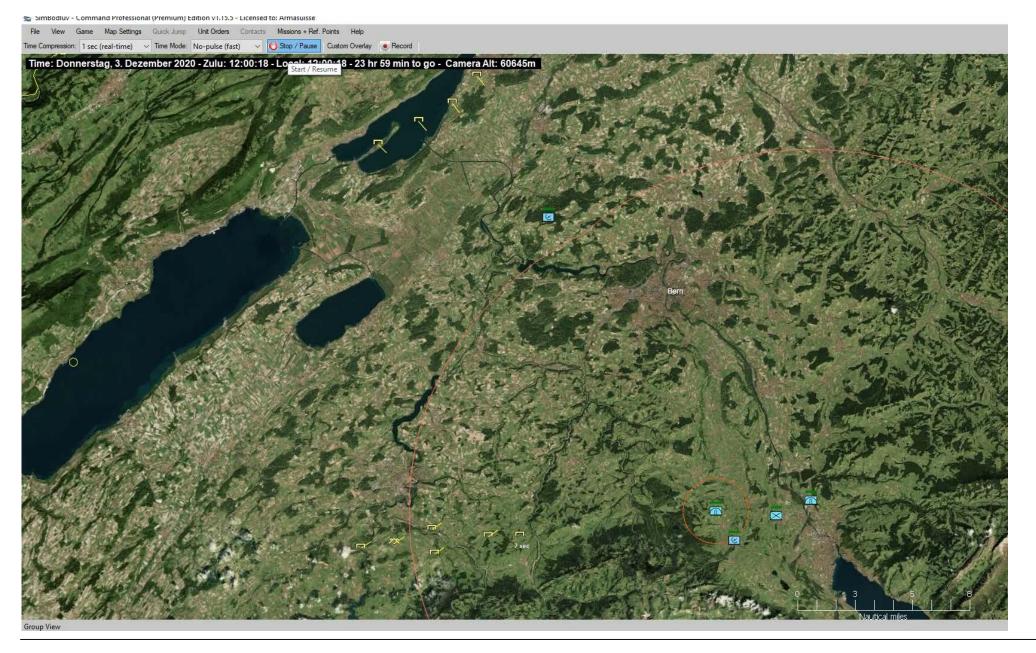
Matthias Sommer, Oleg Szehr

U

Air Defense Scenario



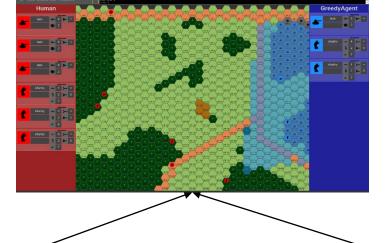




Wargaming: Human - Human

- Training tactical-strategical thinking
- Conflict analysis
- Mission planning
- Validation of known Courses of Action

- Think inside the box
- Slow
- Restricted reproducibility







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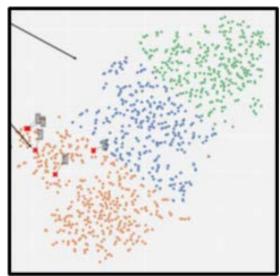
Red Teaming: Human - Al

- Training tactical-strategical thinking
- Conflict analysis
- Mission planning and optimization

 Diverse and new Courses of Action (Think outside of the box)

Slow

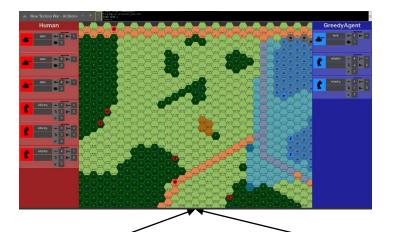






Data Farming: AI - AI

- Concept development and exerpimentation (CD&E)
- Armed Forces development
- Conflict analysis
- Procurement
- Diverse and new Courses of Action (Think outside of the box)
- Fast
- Solution space can be scanned extensively







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O

Al: Reinforcement Learning & Tree Search

 Neural networks are learning by "Trialand-Error"

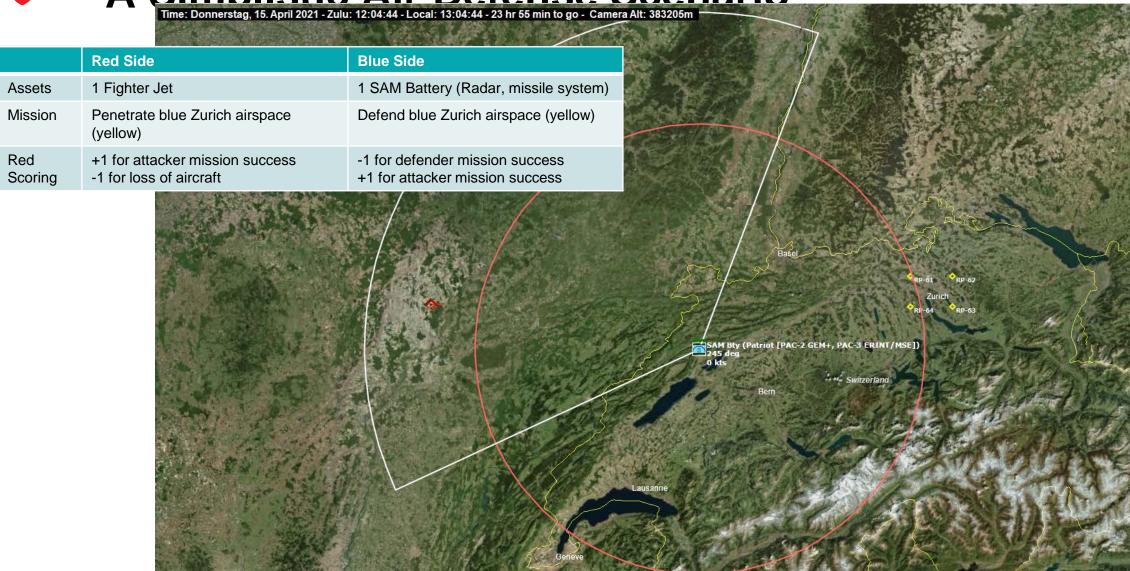








A Simplistic Air Defense Scenario Time: Donnerstag, 15. April 2021 - Zulu: 12:04:44 - Local: 13:04:44 - 23 hr 55 min to go - Camera Alt: 383205m





Discrete Action Scheduling Cycle



Selected action

Propagate 1 time window Include actions from computer role

Environment:

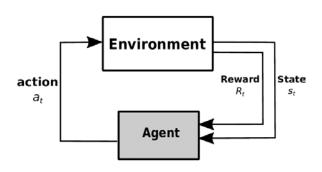
State



Al: Neural Monte Carlo Tree Search (nMCTS)

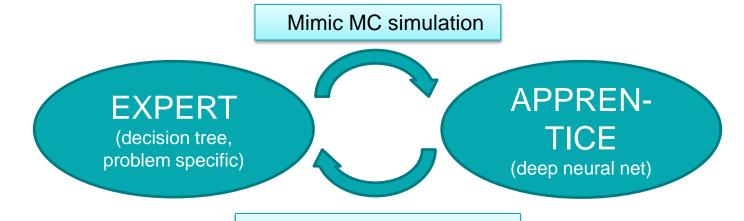
Reinforcement Learning (RL):

through trial, error and successive self-optimization



- Neural RL achieves super-human performance in Go, Chess, Hex ...
- nMCTS combines techniques from
 - Monte Carlo (MC) simulation
 - o game tree search
 - deep learning

nMCTS: Two components, improve one another iteratively



Improve MC probabilities

adding the NN output

EXPERT:

- Run MC simulation
- Record outcomes
 - update quality values of state-action pairs
 - Update MC probabilities
- Perform simulation again, update, ..., ...

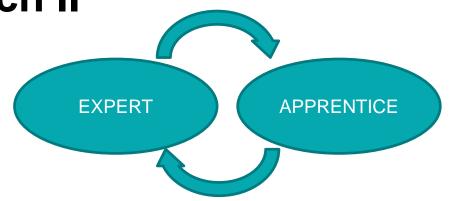
APPRENTICE:

- Train neural network to imitate the behavior of MC
 - Mimic action probabilities (train on tree policy targets)
 - Mimic values of states (train on rewards)

Al: Neural Monte Carlo Tree Search II

EXPERT (MCTS):

- + Provides accurate actions and valuation.
- Requires significant computational effort. Is slow.
- Evaluation of similar states requires full simulation 'from scratch'



APPRENTICE (Neural Network):

- Cannot learn actions/valuation by gradient descent in complex scenarios.
- + Once trained to imitate EXPERT: Fast access to EXPERT advice
- + Generalizes EXPERT advice to similar states.

Each training iteration consists of:

- Multiple neural network-guided MC simulations
- Re-training of the neural network and update of MC probabilities
- Validation to ensure improvement

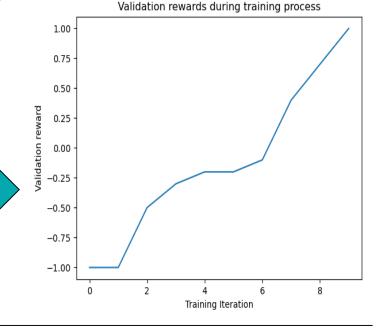
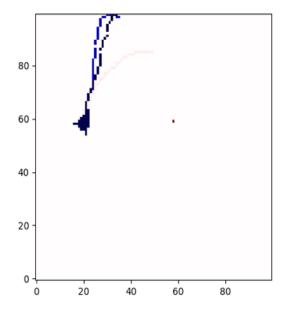


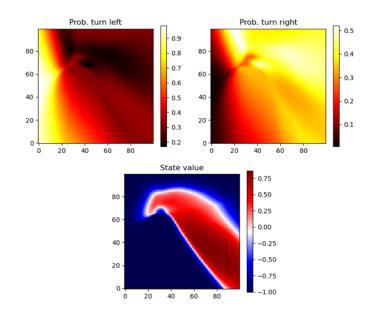


Illustration of training process of AI components:

Expert

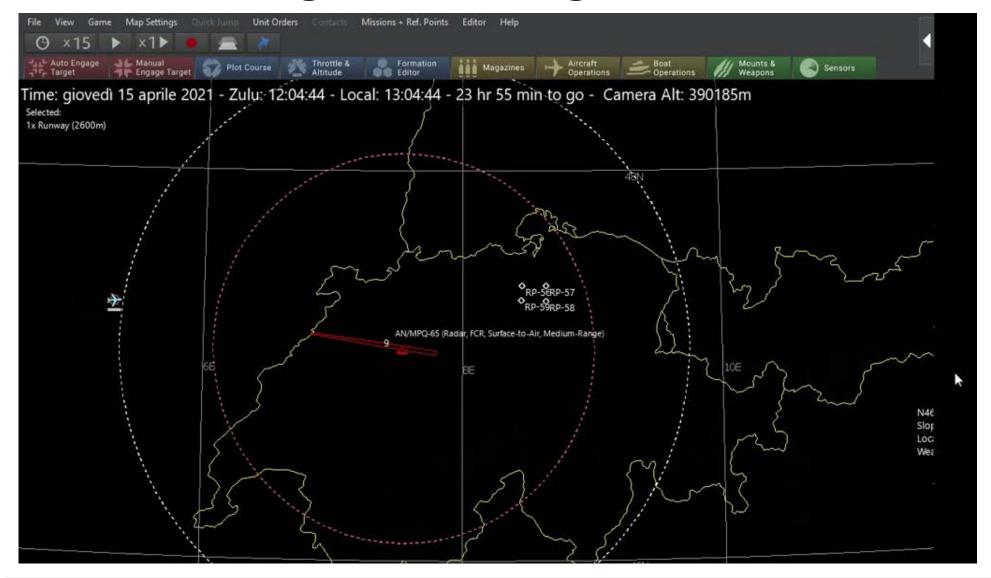


Apprentice



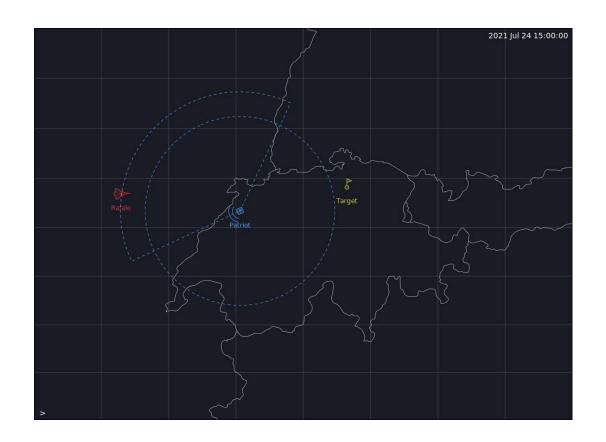
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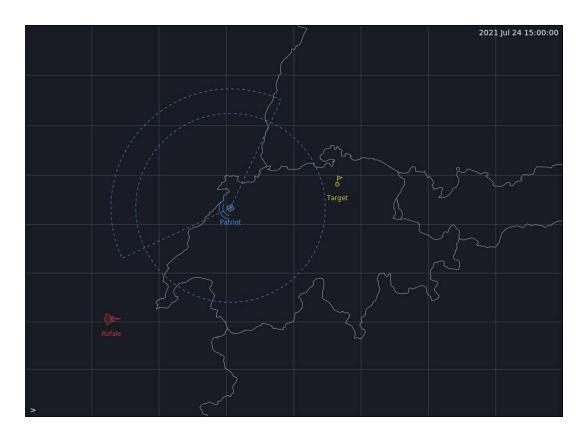
Result: Routing of trained agent





Generalization ability







Conclusions

Proof of Concept

- Successful integration of COTS game with PyTorch
- MCTS able to learn "intelligent behavior"

Bottleneck

- Training performance (API access)
- Workaround: Training with surrogate model

Next step

Invading aircraft can attack air defense battery